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Variational characterization of the Knothe-Rosenblatt type rearrangements and its stochastic version

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1 Introduction.

The Knothe-Rosenblatt rearrangement plays a crucial role in many fields, e.g., the Brunn-Minkowski inequality and statistics (see [12], [13], [22] and the references therein).

Let $d \geq 1$ and let $\mathcal{M}_1(\mathbf{R}^d)$ denote the set of all Borel probability measures on \mathbf{R}^d with a weak topology. For a distribution function F on \mathbf{R} , let

$$F^{-1}(u) := \inf\{x \in \mathbf{R} | u \leq F(x)\} \quad (0 \leq u \leq 1). \quad (1.1)$$

For $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$, $x \in \mathbf{R}$, and $i = 0, 1$, let

$$\begin{aligned} F_{i,k}(x|\mathbf{x}_{k-1}) &:= \begin{cases} P_i((-\infty, x] \times \mathbf{R}^{d-1}) & (k=1), \\ P_i((-\infty, x] \times \mathbf{R}^{d-k} | \mathbf{x}_{k-1}) & (1 < k < d), \\ P_i((-\infty, x] | \mathbf{x}_{d-1}) & (k=d), \end{cases} \\ \varphi_k(\mathbf{x}_k) &:= F_{1,k}(\cdot | \varphi_1(x_1), \dots, \varphi_{k-1}(\mathbf{x}_{k-1}))^{-1}(F_{0,k}(x_k | \mathbf{x}_{k-1})) \quad (1 \leq k \leq d), \end{aligned} \quad (1.2)$$

where $\mathbf{x}_k := (x_i)_{1 \leq i \leq k} \in \mathbf{R}^k$ for $x = (x_i)_{1 \leq i \leq d} \in \mathbf{R}^d$ and $P_i(\cdot | \mathbf{x}_{k-1})$ denotes the regular conditional probability of P_i given \mathbf{x}_{k-1} .

Suppose that $F_{0,k}(\cdot | \mathbf{x}_{k-1})$ is continuous for all $k = 1, \dots, d$. Then P_1 is the image measure of P_0 by

$$T_{KR}(\mathbf{x}_d) := (\varphi_1(x_1), \dots, \varphi_d(\mathbf{x}_d)).$$

T_{KR} is called the **Knothe-Rosenblatt rearrangement**. Suppose, in addition, that $F_{1,k}(\cdot | \mathbf{x}_{k-1})$ is continuous for all $k = 1, \dots, d$. Then T_{KR} is invertible and the minimizer of the following weakly converges to $P_0(dx)\delta_{T_{KR}(x)}(dy)$ as $\varepsilon \rightarrow 0$: for $p > 1$,

$$\inf \left\{ \int_{\mathbf{R}^d \times \mathbf{R}^d} \sum_{k=1}^d \varepsilon^{2(k-1)} |y_k - x_k|^p \mu(dx dy) \middle| \mu(dx \times \mathbf{R}^d) = P_0(dx), \right. \\ \left. \mu(\mathbf{R}^d \times dy) = P_1(dy) \right\}, \quad (1.3)$$

provided $\int_{\mathbf{R}^d} |x|^p (P_0(dx) + P_1(dx))$ is finite (see [2]). Here $\delta_x(dy)$ denotes the delta measure on $\{x\}$.

For $1 \leq k \leq d$, $\mathbf{x}_{k-1} \in \mathbf{R}^{k-1}$, $dF_{0,k}(x|\mathbf{x}_{k-1})\delta_{\varphi_k(\mathbf{x}_k)}(dy)$ is the unique minimizer of

$$\inf \left\{ \int_{\mathbf{R} \times \mathbf{R}} |y - x|^p \mu(dx dy) \middle| \mu(dx \times \mathbf{R}) = dF_{0,k}(x|\mathbf{x}_{k-1}), \right. \\ \left. \mu(\mathbf{R} \times dy) = dF_{1,k}(y|\varphi_1(x_1), \dots, \varphi_{k-1}(\mathbf{x}_{k-1})) \right\} \quad (1.4)$$

(see e.g. [21], [24]). (1.4) also implies that $P_0(d\mathbf{x}_k \times \mathbf{R}^{d-k})\delta_{(\varphi_1(x_1), \dots, \varphi_k(\mathbf{x}_k))}(d\mathbf{y}_k)$ is the unique minimizer of

$$\inf \left\{ \int_{\mathbf{R}^k \times \mathbf{R}^k} |y_k - x_k|^p \mu(dx dy) \middle| \mu(dx \times \mathbf{R}^k) = P_0(dx \times \mathbf{R}^{d-k}), \right. \\ \mu(\mathbf{R}^k \times dy) = P_1(dy \times \mathbf{R}^{d-k}), \\ \left. y_i = \varphi_{i-1}(\mathbf{x}_{i-1}) (i = 1, \dots, k-1), \mu - a.s. \right\}. \quad (1.5)$$

We generalize (1.5) and call the minimizer the **Knothe-Rosenblatt type rearrangement**. We also prove the duality theorem, give the convergence result which generalizes (1.3) by the idea of [2] and consider the similar problems in the stochastic control setting.

2 Knothe-Rosenblatt type rearrangement.

Let $d \geq 2$, $1 \leq d_1 < d$, $c(x, y) : \mathbf{R}^{d-d_1} \times \mathbf{R}^{d-d_1} \mapsto [0, \infty)$ be Borel measurable and $\nu \in \mathcal{M}_1(\mathbf{R}^{2d_1})$. For $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$, let

$$T(P_0, P_1|\nu) := \inf \left\{ \int_{\mathbf{R}^d \times \mathbf{R}^d} c(\mathbf{x}_{d_1,d}, \mathbf{y}_{d_1,d}) \mu(dx dy) \middle| \right. \\ \mu(d\mathbf{x}_{d_1} \times \mathbf{R}^{d-d_1} \times d\mathbf{y}_{d_1} \times \mathbf{R}^{d-d_1}) = \nu(d\mathbf{x}_{d_1} d\mathbf{y}_{d_1}), \\ \left. \mu(dx \times \mathbf{R}^d) = P_0(dx), \mu(\mathbf{R}^d \times dy) = P_1(dy) \right\}, \quad (2.1)$$

where $\mathbf{x}_{i,j} := (x_k)_{i+1 \leq k \leq j} \in \mathbf{R}^{j-i}$ for $x = (x_k)_{1 \leq k \leq d} \in \mathbf{R}^d$. If the set over which the infimum is taken is empty, then we consider the infimum is equal to infinity. If there exists a Borel measurable function $\varphi : \mathbf{R}^{d_1} \mapsto \mathbf{R}^{d_1}$ such that $\mathbf{y}_{d_1} = \varphi(\mathbf{x}_{d_1})$, ν -a.s., then we write, for simplicity,

$$T(P_0, P_1 | \varphi) := T(P_0, P_1 | \nu).$$

We first show the existence of the Knothe-Rosenblatt type rearrangement.

Proposition 2.1 *Suppose that c is lower semi-continuous. Then, for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$, $T(P_0, P_1 | \nu)$ has a minimizer, provided it is finite.*

(Proof) Let $\{\mu_n\}_{n \geq 1}$ be a minimizing sequence of $T(P_0, P_1 | \nu)$. Since $\mu_n(dx \times \mathbf{R}^d) = P_0(dx)$ and $\mu_n(\mathbf{R}^d \times dy) = P_1(dy)$, it has a weakly convergent subsequence which we denote by $\{\mu_{n(k)}\}_{k \geq 1}$. Let μ denote the limit. Then by Skorohod's representation theorem, Fatou's lemma and the lower semicontinuity of c ,

$$\begin{aligned} T(P_0, P_1 | \nu) &= \lim_{k \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} c(\mathbf{x}_{d_1, d}, \mathbf{y}_{d_1, d}) \mu_{n(k)}(dx dy) \\ &\geq \int_{\mathbf{R}^d \times \mathbf{R}^d} c(\mathbf{x}_{d_1, d}, \mathbf{y}_{d_1, d}) \mu(dx dy). \end{aligned} \quad (2.2)$$

For any $f \in C(\mathbf{R}^{d_1} \times \mathbf{R}^{d_1})$,

$$\begin{aligned} \int_{\mathbf{R}^d \times \mathbf{R}^d} f(\mathbf{x}_{d_1}, \mathbf{y}_{d_1}) \mu(dx dy) &= \lim_{k \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} f(\mathbf{x}_{d_1}, \mathbf{y}_{d_1}) \mu_{n(k)}(dx dy) \\ &= \int_{\mathbf{R}^{d_1} \times \mathbf{R}^{d_1}} f(\mathbf{x}_{d_1}, \mathbf{y}_{d_1}) \nu(d\mathbf{x}_{d_1} d\mathbf{y}_{d_1}). \end{aligned} \quad (2.3)$$

In the same way, one can show that $\mu(dx \times \mathbf{R}^d) = P_0(dx)$ and $\mu(\mathbf{R}^d \times dy) = P_1(dy)$. \square

2.1 Duality Theorem

It is easy to see that the following holds:

$$\begin{aligned} T(P_0, P_1 | \varphi) &= \inf \left\{ \int_{\mathbf{R}^d \times \mathbf{R}^d} \frac{c(\mathbf{x}_{d_1, d}, \mathbf{y}_{d_1, d})}{1_{\{\varphi(\mathbf{x}_{d_1})\}}(\mathbf{y}_{d_1})} \mu(dx dy) \right. \\ &\quad \left. \mu(dx \times \mathbf{R}^d) = P_0(dx), \mu(\mathbf{R}^d \times dy) = P_1(dy) \right\}, \end{aligned} \quad (2.4)$$

where $1_A(x) := 1$ if $x \in A$ and $:= 0$ if $x \notin A$ for the set A . This leads us to the duality theorem for $T(P_0, P_1 | \varphi)$ which can be obtained from [11] (see also p. 76 in Vol. 1 of [21]).

Theorem 2.1 For any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,

$$T(P_0, P_1|\varphi) = \sup \left\{ \int_{\mathbf{R}^d} f_1(y) P_1(dy) - \int_{\mathbf{R}^d} f_0(x) P_0(dx) \mid f_0, f_1 \in C_b(\mathbf{R}^d), \right. \\ \left. f_1(y) - f_0(x) \leq \frac{c(\mathbf{x}_{d_1,d}, \mathbf{y}_{d_1,d})}{1_{\{\varphi(\mathbf{x}_{d_1})\}}(\mathbf{y}_{d_1})} \right\}. \quad (2.5)$$

For $f \in C_b(\mathbf{R}^d)$ and $x = (\mathbf{x}_{d_1}, \mathbf{x}_{d_1,d}) \in \mathbf{R}^d$,

$$v(x; f|\varphi) := \sup \{ f(\varphi(\mathbf{x}_{d_1}), y) - c(\mathbf{x}_{d_1,d}, y) \mid y \in \mathbf{R}^{d-d_1} \}. \quad (2.6)$$

Then, from (2.5),

$$T(P_0, P_1|\varphi) = \sup \left\{ \int_{\mathbf{R}^d} f(y) P_1(dy) - \int_{\mathbf{R}^d} v(x; f|\varphi) P_0(dx) \mid f \in C_b(\mathbf{R}^d) \right\}. \quad (2.7)$$

We easily obtain the following (see e.g. (2.8)-(2.9) in [16]).

Proposition 2.2 Suppose that $\varphi \in C(\mathbf{R}^{d_1} : \mathbf{R}^{d_1})$, $c(x, y) \in C(\mathbf{R}^{d-d_1} \times \mathbf{R}^{d-d_1} : [0, \infty))$ and $\lim_{|y-x| \rightarrow \infty} c(x, y) = \infty$. Then for any $f \in C_b(\mathbf{R}^d)$, $v(\cdot; f|\varphi)$ is continuous.

We formally derive the Hamilton-Jacobi Equation (HJ Eqn for short) for $v(x; f|\varphi)$.
Let

$$\begin{aligned} \Phi(t, x) &:= x + t(\varphi(x) - x), \\ b(t, x) &:= \varphi(\Phi(t, \cdot)^{-1}(x)) - \Phi(t, \cdot)^{-1}(x) \quad ((t, x) \in [0, 1] \times \mathbf{R}^{d_1}), \end{aligned} \quad (2.8)$$

provided it exists. Then

$$\frac{d\Phi(t, x)}{dt} = \varphi(x) - x = b(t, \Phi(t, x)). \quad (2.9)$$

In case $c(x, y) = \ell(y - x)$ for a convex ℓ , we consider the following HJ Eqn:

$$\frac{\partial v(t, x)}{\partial t} + \langle \nabla_{d_1} v(t, x), b(t, \mathbf{x}_{d_1}) \rangle + h(\nabla_{d_1,d} v(t, x)) = 0 \quad ((t, x) \in (0, 1) \times \mathbf{R}^d), \quad (2.10)$$

where $\nabla_{d_1} := (\partial/\partial x_i)_{i=1}^{d_1}$, $\nabla_{d_1,d} := (\partial/\partial x_i)_{i=d_1+1}^d$ and

$$h(z) := \sup \{ \langle u, z \rangle - \ell(u) \mid u \in \mathbf{R}^{d-d_1} \} \quad (z \in \mathbf{R}^{d-d_1}).$$

Then we have

Proposition 2.3 Suppose that $c(x, y) = \ell(y - x)$ for a convex ℓ , that $\Phi(t, \cdot)$ is injective for all $t \in [0, 1]$, that the HJ Eqn (2.10) has a classical solution v and that the following ODE has an absolutely continuous solution: for any $\phi_2(0) = \mathbf{x}_{d_1, d}$

$$\frac{d\phi_2(t)}{dt} = \nabla h(\nabla_{\mathbf{x}_{d_1, d}} v(t, \Phi(t, \mathbf{x}_{d_1}), \phi_2(t))). \quad (2.11)$$

Then $v(0, x) = v(x; v(1, \cdot)|\varphi)$.

(Proof) For any $\phi_2 \in AC(\mathbf{R}^{d-d_1})$, from (2.9), we have

$$\begin{aligned} & v(1, \Phi(1, \mathbf{x}_{d_1}), \phi_2(1)) - v(0, \Phi(0, \mathbf{x}_{d_1}), \phi_2(0)) \\ &= \int_0^1 \left\{ \frac{\partial v(t, \Phi(t, \mathbf{x}_{d_1}), \phi_2(t))}{\partial t} + \langle \nabla_{d_1} v(t, \Phi(t, \mathbf{x}_{d_1}), \phi_2(t)), b(t, \Phi(t, \mathbf{x}_{d_1})) \rangle \right. \\ &\quad \left. + \langle \nabla_{d_1, d} v(t, \Phi(t, \mathbf{x}_{d_1}), \phi_2(t)), \frac{d\phi_2(t)}{dt} \rangle \right\} dt \\ &= \int_0^1 \left\{ -h(\nabla_{d_1, d} v(t, \Phi(t, \mathbf{x}_{d_1}), \phi_2(t))) \right. \\ &\quad \left. + \langle \nabla_{d_1, d} v(t, \Phi(t, \mathbf{x}_{d_1}), \phi_2(t)), \frac{d\phi_2(t)}{dt} \rangle \right\} dt \\ &\leq \int_0^1 \ell\left(\frac{d\phi_2(t)}{dt}\right) dt, \end{aligned} \quad (2.12)$$

where the equality holds if (2.11) holds. By Jensen's inequality,

$$\begin{aligned} v(0, x) &= \sup \left\{ v(1, \varphi(\mathbf{x}_{d_1}), \phi_2(1)) - \int_0^1 \ell\left(\frac{d\phi_2(t)}{dt}\right) dt \mid \phi_2(0) = \mathbf{x}_{d_1, d} \right\} \\ &= \sup \left\{ v(1, \varphi(\mathbf{x}_{d_1}), \phi_2(1)) - \ell(\phi_2(1) - \phi_2(0)) \mid \phi_2(0) = \mathbf{x}_{d_1, d} \right\} \\ &= v(x; v(1, \cdot)|\varphi). \square \end{aligned} \quad (2.13)$$

Before we formulate the duality theorem in the framework of the theory of viscosity solutions, we give assumptions.

(A.1). $b(t, x)$ is bounded and there exists $K > 0$ such that

$$|b(t, x) - b(t, y)| \leq K|x - y| \quad (t \in [0, 1], x, y \in \mathbf{R}^{d_1}).$$

(A.2). There exists $m \in C([0, 1] \times \mathbf{R}^{d_1} \times [0, 1] \times \mathbf{R}^{d_1} \times [0, \infty))$ such that $m(t, x, s, y, 0) = 0$ and that

$$|b(t, x) - b(s, y)| \leq m(t, x, s, y, |t - s| + |x - y|) \quad (t, s \in [0, 1], x, y \in \mathbf{R}^{d_1}).$$

(A.3). $\ell : \mathbf{R}^{d-d_1} \mapsto [0, \infty)$ is convex and $\liminf_{|v| \rightarrow \infty} \frac{\ell(v)}{|v|} = \infty$.

Example 2.1 Suppose that $d_1 = 1$. Then (A.1)-(A.2) holds if $1 < d\varphi(x)/dx \leq K + 1$.

For $(t, x) \in [0, 1] \times \mathbf{R}^d$ and $f \in C_b(\mathbf{R}^d)$,

$$\begin{aligned} & v(t, x; f|\varphi) \\ &:= \sup \left\{ f(\Phi(1, \mathbf{y}_{d_1}), \phi_2(1)) - \int_t^1 \ell \left(\frac{d\phi_2(s)}{ds} \right) ds \mid (\Phi(t, \mathbf{y}_{d_1}), \phi_2(t)) = x \right\}. \end{aligned} \quad (2.14)$$

Then it is easy to see that the following holds:

$$\begin{aligned} & v(t, x; f|\varphi) \\ &= \sup \left\{ f(\mathbf{x}_{d_1} + (1-t)b(t, \mathbf{x}_{d_1}), y) - (1-t)\ell \left(\frac{y - \mathbf{x}_{d_1,d}}{1-t} \right) \mid y \in \mathbf{R}^{d-d_1} \right\}. \end{aligned} \quad (2.15)$$

(see (2.8)). We also have

Corollary 2.1 Suppose that $c(x, y) = \ell(y - x)$ and that (A.1)-(A.3) hold. Then for any Lipschitz continuous $f : \mathbf{R}^d \mapsto \mathbf{R}$, $v(t, x; f|\varphi)$ is a Lipschitz continuous viscosity solution of (2.10). In particular, for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,

$$T(P_0, P_1|\varphi) = \sup \left\{ \int_{\mathbf{R}^d} v(1, y) P_1(dy) - \int_{\mathbf{R}^d} v(0, x) P_0(dx) \mid v(1, \cdot) \in C_b^\infty(\mathbf{R}^d) \right\}, \quad (2.16)$$

where $v(t, x)$ denotes a bounded uniformly continuous viscosity solution of (2.10).

(Proof) In the same way as in p. 127 in [4], by (A.1) and (A.3), one can prove that $v(\cdot, \cdot; f|\varphi)$ is Lipschitz continuous for Lipschitz continuous $f : \mathbf{R}^d \mapsto \mathbf{R}$. In addition, from Chap. II.16 of [7], under (A.1)-(A.3), $v(t, x; f|\varphi)$ is a bounded, uniformly continuous viscosity solution of (2.10). It is easy to see that the supremum in (2.7) can be taken only over all $f \in C_b^\infty(\mathbf{R}^d)$. For $n \geq 1$, $f \in C_b^\infty(\mathbf{R}^d)$ and $(t, x) \in [0, 1] \times \mathbf{R}^d$,

$$\begin{aligned} v_n(t, x; f) &:= \sup \left\{ f(\mathbf{x}_{d_1} + (1-t)b(t, \mathbf{x}_{d_1}), \phi_2(1)) - \int_t^1 \ell \left(\frac{d\phi_2(s)}{ds} \right) ds \mid \right. \\ &\quad \left. \phi_2(t) = \mathbf{x}_{d_1,d}, \left| \frac{d\phi_2(s)}{ds} \right| \leq n \right\}. \end{aligned} \quad (2.17)$$

Then, from Theorem 10.1 in p. 95 of [7], under (A.1), $v_n(t, x; f)$ is the unique bounded uniformly continuous viscosity solution of the following HJ Eqn: for $(t, x) \in (0, 1) \times \mathbf{R}^d$,

$$\begin{aligned} \frac{\partial v(t, x)}{\partial t} + \langle \nabla_{d_1} v(t, x), b(t, \mathbf{x}_{d_1}) \rangle + h_n(\nabla_{d_1, d} v(t, x)) &= 0, \\ v(1, x) &= f(x), \end{aligned} \quad (2.18)$$

where

$$h_n(z) := \sup\{\langle u, z \rangle - \ell(u) \mid u \in \mathbf{R}^{d-d_1}, |u| \leq n\}.$$

Let \bar{v} be a bounded uniformly continuous viscosity solution of (2.10) with $\bar{v}(1, x) = f(x)$. Then it is a bounded uniformly continuous viscosity supersolution of (2.18) with $\bar{v}(1, x) = f(x)$ and

$$v_n(t, x; f) \leq \bar{v}(t, x) \quad (2.19)$$

from Theorem 9.1 in p. 86 of [7]. Let $n \rightarrow \infty$ in (2.19). Then we obtain $v(t, x; f|\varphi) \leq \bar{v}(t, x) \square$.

2.2 Convergence Theorem

Let $2 \leq k \leq d$, $0 = d_0 < d_1 < \dots < d_k = d$ and

(A.4) $c_i \in LSC(\mathbf{R}^{d_i-d_{i-1}} \times \mathbf{R}^{d_i-d_{i-1}} : [0, \infty))$ ($i = 1, \dots, k$).

For $\varepsilon \geq 0$, $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,

$$\begin{aligned} T^\varepsilon(P_0, P_1) &:= \inf \left\{ \int_{\mathbf{R}^d \times \mathbf{R}^d} \sum_{i=1}^k \varepsilon^{i-1} c_i(\mathbf{x}_{d_{i-1}, d_i}, \mathbf{y}_{d_{i-1}, d_i}) \mu(d\mathbf{x} d\mathbf{y}) \right. \\ &\quad \left. \mu(d\mathbf{x} \times \mathbf{R}^d) = P_0(d\mathbf{x}), \mu(\mathbf{R}^d \times d\mathbf{y}) = P_1(d\mathbf{y}) \right\}. \end{aligned} \quad (2.20)$$

It is known that if $c_i(x, y) = \ell_i(y - x)$ and ℓ_i is strictly convex and superlinear ($i = 1, \dots, k$) and if $P_0(dx)$ is absolutely continuous with respect to the Lebesgue measure dx , then $T^\varepsilon(P_0, P_1)$ has the unique minimizer, provided that it is finite (see e.g. [21], [24], [25]).

$$\begin{aligned} T_1(P_{0,1}, P_{1,1}) &:= \inf \left\{ \int_{\mathbf{R}^{d_1} \times \mathbf{R}^{d_1}} c_1(x, y) \mu(d\mathbf{x} d\mathbf{y}) \right. \\ &\quad \left. \mu(d\mathbf{x} \times \mathbf{R}^{d_1}) = P_{0,1}(d\mathbf{x}), \mu(\mathbf{R}^{d_1} \times d\mathbf{y}) = P_{1,1}(d\mathbf{y}) \right\}, \end{aligned} \quad (2.21)$$

where $P_{t,i}(d\mathbf{x}_{d_i}) := P_t(d\mathbf{x}_{d_i} \times \mathbf{R}^{d-d_i})$ ($t = 0, 1$). For $i \geq 2$ and $\nu_{i-1} \in \mathcal{M}_1(\mathbf{R}^{2d_{i-1}})$,

$$T_i(P_{0,i}, P_{1,i}|\nu_{i-1}) := \inf \left\{ \int_{\mathbf{R}^{d_i} \times \mathbf{R}^{d_i}} c_i(\mathbf{x}_{d_{i-1}, d_i}, \mathbf{y}_{d_{i-1}, d_i}) \mu(dx dy) \right| \\ \mu(d\mathbf{x}_{d_{i-1}} \times \mathbf{R}^{d_i-d_{i-1}} \times d\mathbf{y}_{d_{i-1}} \times \mathbf{R}^{d_i-d_{i-1}}) = \nu_{i-1}(d\mathbf{x}_{d_{i-1}} d\mathbf{y}_{d_{i-1}}), \\ \mu(dx \times \mathbf{R}^{d_i}) = P_{0,i}(dx), \mu(\mathbf{R}^{d_i} \times dy) = P_{1,i}(dy) \Big\}. \quad (2.22)$$

The following theorem can be proved in the same way as [2] (see also section 1) and is proved for the readers' convenience.

Theorem 2.2 *Let $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$. Suppose that $k = 2$ and (A.4) holds and that $T_1(P_{0,1}, P_{1,1})$ and $T_2(P_0, P_1|\nu_1)$ have the unique minimizers ν_1 and ν_2 , respectively. Then a minimizer of $T^\varepsilon(P_0, P_1)$ exists and weakly converges to ν_2 as $\varepsilon \rightarrow 0$ and the following holds:*

$$\lim_{\varepsilon \rightarrow 0} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_1(\mathbf{x}_{d_1}, \mathbf{y}_{d_1}) \mu^\varepsilon(dx dy) = T_1(P_{0,1}, P_{1,1}), \quad (2.23)$$

$$\lim_{\varepsilon \rightarrow 0} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_2(\mathbf{x}_{d_1, d}, \mathbf{y}_{d_1, d}) \mu^\varepsilon(dx dy) = T_2(P_0, P_1|\nu_1). \quad (2.24)$$

(Proof). In the same way as in the proof of Proposition 2.1, by a standard method, one can show that $T^\varepsilon(P_0, P_1)$ has a minimizer μ^ε , since

$$T^\varepsilon(P_0, P_1) \leq T_1(P_{0,1}, P_{1,1}) + \varepsilon T_2(P_0, P_1|\nu_1) < +\infty. \quad (2.25)$$

Since the set of μ for which $\mu(dx \times \mathbf{R}^d) = P_0(dx)$ and $\mu(\mathbf{R}^d \times dy) = P_1(dy)$ is compact, any sequence $\{\mu^{\varepsilon_n}\}_{n \geq 1}$ ($\varepsilon_n \rightarrow 0$ as $n \rightarrow \infty$) has a weakly convergent subsequence $\{\mu^{\varepsilon_{n(\ell)}}\}_{\ell \geq 1}$ and for the limit μ ,

$$\mu_1(d\mathbf{x}_{d_1} d\mathbf{y}_{d_1}) := \mu(d\mathbf{x}_{d_1} \times \mathbf{R}^{d-d_1} \times d\mathbf{y}_{d_1} \times \mathbf{R}^{d-d_1})$$

is the minimizer of $T_1(P_{0,1}, P_{1,1})$ by the uniqueness of the minimizer and (2.23) holds. Indeed, from (2.25),

$$\begin{aligned} T_1(P_{0,1}, P_{1,1}) &\leq \int_{\mathbf{R}^{d_1} \times \mathbf{R}^{d_1}} c_1(x, y) \mu_1(dx dy) = \int_{\mathbf{R}^d \times \mathbf{R}^d} c_1(\mathbf{x}_{d_1}, \mathbf{y}_{d_1}) \mu(dx dy) \\ &\leq \liminf_{\ell \rightarrow \infty} T^{\varepsilon_{n(\ell)}}(P_0, P_1) \leq \limsup_{\ell \rightarrow \infty} T^{\varepsilon_{n(\ell)}}(P_0, P_1) \\ &\leq T_1(P_{0,1}, P_{1,1}). \end{aligned} \quad (2.26)$$

Since

$$T_1(P_{0,1}, P_{1,1}) + \varepsilon \int_{\mathbf{R}^d \times \mathbf{R}^d} c_2(\mathbf{x}_{d_1,d}, \mathbf{y}_{d_1,d}) \mu^\varepsilon(d\mathbf{x}d\mathbf{y}) \leq T^\varepsilon(P_0, P_1), \quad (2.27)$$

we also have, from (2.25) and (2.27),

$$\begin{aligned} T_2(P_0, P_1 | \nu_1) &\leq \int_{\mathbf{R}^d \times \mathbf{R}^d} c_2(\mathbf{x}_{d_1,d}, \mathbf{y}_{d_1,d}) \mu(d\mathbf{x}d\mathbf{y}) \\ &\leq \liminf_{\ell \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_2(\mathbf{x}_{d_1,d}, \mathbf{y}_{d_1,d}) \mu^{\varepsilon_{n(\ell)}}(d\mathbf{x}d\mathbf{y}) \\ &\leq \limsup_{\ell \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_2(\mathbf{x}_{d_1,d}, \mathbf{y}_{d_1,d}) \mu^{\varepsilon_{n(\ell)}}(d\mathbf{x}d\mathbf{y}) \\ &\leq T_2(P_0, P_1 | \nu_1). \end{aligned} \quad (2.28)$$

The uniqueness of the minimizer of $T_2(P_0, P_1 | \nu_1)$ completes the proof. \square

Theorem 2.3 *Let $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$. Suppose that (A.4) holds, that $T_1(P_{0,1}, P_{1,1})$ and $T_i(P_{0,i}, P_{1,i} | \nu_{i-1})$ have the unique minimizers ν_1 and ν_i ($i = 2, \dots, k$), respectively and that $\nu \mapsto T_i(P_{0,i}, P_{1,i} | \nu)$ is continuous ($i = 3, \dots, k$). Then a minimizer of $T^\varepsilon(P_0, P_1)$ exists and weakly converges to ν_k as $\varepsilon \rightarrow 0$ and the following holds:*

$$\lim_{\varepsilon \rightarrow 0} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_1(\mathbf{x}_{d_1}, \mathbf{y}_{d_1}) \mu^\varepsilon(d\mathbf{x}d\mathbf{y}) = T_1(P_{0,1}, P_{1,1}), \quad (2.29)$$

$$\lim_{\varepsilon \rightarrow 0} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_i(\mathbf{x}_{d_{i-1},d_i}, \mathbf{y}_{d_{i-1},d_i}) \mu^\varepsilon(d\mathbf{x}d\mathbf{y}) = T_i(P_{0,i}, P_{1,i} | \nu_{i-1}) \quad (i = 2, \dots, k). \quad (2.30)$$

(Proof). In the same way as in (2.25), one can show that $T^\varepsilon(P_0, P_1)$ has a minimizer μ^ε and that any subsequence $\{\mu^{\varepsilon_n}\}_{n \geq 1}$ ($\varepsilon_n \rightarrow 0$ as $n \rightarrow \infty$) has a weakly convergent subsequence $\{\mu^{\varepsilon_{n(\ell)}}\}_{\ell \geq 1}$. Let μ denote the weak limit of $\mu^{\varepsilon_{n(\ell)}}$ as $\ell \rightarrow \infty$. We prove the theorem by induction. For $i = 2, \dots, k$,

$$\begin{aligned} &T_{i-1}^\varepsilon(P_{0,i-1}, P_{1,i-1}) \\ &:= \inf \left\{ \int_{\mathbf{R}^{d_{i-1}} \times \mathbf{R}^{d_{i-1}}} \sum_{j=1}^{i-1} \varepsilon^{j-1} c_j(\mathbf{x}_{d_{j-1},d_j}, \mathbf{y}_{d_{j-1},d_j}) \nu(d\mathbf{x}d\mathbf{y}) \right. \\ &\quad \left. \nu(d\mathbf{x} \times \mathbf{R}^{d_{i-1}}) = P_{0,i-1}(d\mathbf{x}), \nu(\mathbf{R}^{d_{i-1}} \times d\mathbf{y}) = P_{1,i-1}(d\mathbf{y}) \right\}. \end{aligned} \quad (2.31)$$

Let μ_{i-1}^ε and $\nu_{i,j}^\varepsilon$ denote a minimizer of $T_{i-1}^\varepsilon(P_{0,i-1}, P_{1,i-1})$ and $T_j(P_{0,j}, P_{1,j} | \nu_{i,j-1}^\varepsilon)$ ($j = i, \dots, k$), respectively, where $\nu_{i,i-1}^\varepsilon := \mu_{i-1}^\varepsilon$. Then

$$\begin{aligned}
& T_{i-1}^\varepsilon(P_{0,i-1}, P_{1,i-1}) + \int_{\mathbf{R}^d \times \mathbf{R}^d} \sum_{j=i}^k \varepsilon^{j-1} c_j(\mathbf{x}_{d_{j-1}, d_j}, \mathbf{y}_{d_{j-1}, d_j}) \mu^\varepsilon(d\mathbf{x}d\mathbf{y}) \\
& \leq T^\varepsilon(P_0, P_1) \\
& \leq T_{i-1}^\varepsilon(P_{0,i-1}, P_{1,i-1}) + \sum_{j=i}^k \varepsilon^{j-1} T_j(P_{0,j}, P_{1,j} | \nu_{i,j-1}^\varepsilon).
\end{aligned} \tag{2.32}$$

From Theorem 2.2, $\mu_2^\varepsilon \rightarrow \nu_2$ as $\varepsilon \rightarrow 0$ and (2.23)-(2.24) holds. Suppose that $\mu_i^\varepsilon \rightarrow \nu_i$ as $\varepsilon \rightarrow 0$ for $i \leq k-1$. In the same way as in Theorem 2.2, one can show that for $j = 1, 2$,

$$\mu(d\mathbf{x}_{d_j} \times \mathbf{R}^{d-d_j} \times d\mathbf{y}_{d_j} \times \mathbf{R}^{d-d_j}) = \nu_j(d\mathbf{x}_{d_j} d\mathbf{y}_{d_j}). \tag{2.33}$$

Suppose that (2.33) holds for $j = i-1$. Then, from (2.32) and the assumption of induction,

$$\begin{aligned}
T_i(P_{0,i}, P_{1,i} | \nu_{i-1}) & \leq \int_{\mathbf{R}^d \times \mathbf{R}^d} c_i(\mathbf{x}_{d_{i-1}, d_i}, \mathbf{y}_{d_{i-1}, d_i}) \mu(d\mathbf{x}d\mathbf{y}) \\
& \leq \liminf_{\ell \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_i(\mathbf{x}_{d_{i-1}, d_i}, \mathbf{y}_{d_{i-1}, d_i}) \mu^{\varepsilon_n(\ell)}(d\mathbf{x}d\mathbf{y}) \\
& \leq \limsup_{\ell \rightarrow \infty} \int_{\mathbf{R}^d \times \mathbf{R}^d} c_i(\mathbf{x}_{d_{i-1}, d_i}, \mathbf{y}_{d_{i-1}, d_i}) \mu^{\varepsilon_n(\ell)}(d\mathbf{x}d\mathbf{y}) \\
& \leq \lim_{\ell \rightarrow \infty} T_i(P_{0,i}, P_{1,i} | \mu_{i-1}^{\varepsilon_n(\ell)}) = T_i(P_0, P_1 | \nu_{i-1}).
\end{aligned} \tag{2.34}$$

(2.34) implies (2.30) and the uniqueness of the minimizer ν_i of $T_i(P_{0,i}, P_{1,i} | \nu_{i-1})$ implies that (2.33) holds for $j = i$. \square

From (2.32), we also have

Proposition 2.4 *Suppose that the assumption in Theorem 2.3 holds. Then, for $i = 1, \dots, k-1$,*

$$0 \leq \frac{\int_{\mathbf{R}^d \times \mathbf{R}^d} \sum_{j=1}^i \varepsilon^{j-1} c_j(\mathbf{x}_{d_{j-1}, d_j}, \mathbf{y}_{d_{j-1}, d_j}) \mu^\varepsilon(d\mathbf{x}d\mathbf{y}) - T_i^\varepsilon(P_{0,i}, P_{1,i})}{\varepsilon^i} \rightarrow 0 \quad (\varepsilon \rightarrow 0). \tag{2.35}$$

We don't know the real convergence rate of (2.35).

Example 2.2 *Let $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$. Suppose that*

(i) $d_{i+1} = d_i + 1$ ($i = 1, \dots, k-1$),

(ii) $c_i(x, y) = \ell_i(y - x)$ and $\ell_i : \mathbf{R}^{d_i} \mapsto [0, \infty)$ is strictly convex and superlinear ($i = 1, \dots, k$),

(iii) P_0 is absolutely continuous with respect to the Lebesgue measure dx ,

(iv) $T_1(P_{0,1}, P_{1,1})$ is finite.

Then $T_1(P_{0,1}, P_{1,1})$ has the unique minimizer ν_1 which can be written as follows:

$$\nu_1(d\mathbf{x}_{d_1} d\mathbf{y}_{d_1}) = P_{0,1}(d\mathbf{x}_{d_1}) \delta_{\phi_1(\mathbf{x}_{d_1})}(d\mathbf{y}_{d_1}), \quad (2.36)$$

where ϕ_1 is a Borel measurable function (see e.g. [21], [24]).

Suppose, in addition, that

(v) $T_i(P_{0,i}, P_{1,i}|\nu_{i-1})$ is finite for $i = 2, \dots, k$. (If $T_i(P_{0,i}, P_{1,i}|\nu_{i-1})$ is finite, then it has a minimizer (see the proof of Prop. 2.1).)

Then the following holds:

$$\nu_i(d\mathbf{x}_{d_i} d\mathbf{y}_{d_i}) = P_{0,i}(d\mathbf{x}_{d_i}) \delta_{\Phi_{\nu_0, \dots, \nu_{i-1}}(\mathbf{x}_{d_i})}(d\mathbf{y}_{d_i}), \quad (2.37)$$

where $\Phi_{\nu_0, \dots, \nu_{i-1}}(\mathbf{x}_{d_i}) := (\phi_{\nu_0}(\mathbf{x}_{d_1}), \dots, \phi_{\nu_{i-1}}(\mathbf{x}_{d_{i-1}}))$, $\phi_{\nu_0} := \phi_1$ and

$$\begin{aligned} \phi_{\nu_{i-1}}(\mathbf{x}_{d_i}) &:= (F_{\nu_{i-1},1}(\cdot|\mathbf{x}_{d_{i-1}}, \Phi_{\nu_0, \dots, \nu_{i-2}}(\mathbf{x}_{d_{i-1}})))^{-1}(F_{\nu_{i-1},0}(x_{d_i}|\mathbf{x}_{d_{i-1}}, \Phi_{\nu_0, \dots, \nu_{i-2}}(\mathbf{x}_{d_{i-1}}))), \\ &\quad F_{\nu_{i-1},1}(x|\mathbf{x}_{d_{i-1}}, \Phi_{\nu_0, \dots, \nu_{i-2}}(\mathbf{x}_{d_{i-1}})) \\ &:= \nu_{i-1}(\mathbf{R} \times (-\infty, x] | (\mathbf{x}_{d_{i-1}}, \Phi_{\nu_0, \dots, \nu_{i-2}}(\mathbf{x}_{d_{i-1}}))), \\ &\quad F_{\nu_{i-1},0}(x_{d_i}|\mathbf{x}_{d_{i-1}}, \Phi_{\nu_0, \dots, \nu_{i-2}}(\mathbf{x}_{d_{i-1}})) \\ &:= \nu_{i-1}((-\infty, x] \times \mathbf{R} | (\mathbf{x}_{d_{i-1}}, \Phi_{\nu_0, \dots, \nu_{i-2}}(\mathbf{x}_{d_{i-1}}))). \end{aligned}$$

In particular, $\phi_{\nu_{i-1}}$ is a minimizer of the following:

$$\min \left\{ \int_{\mathbf{R}^{d_i}} \ell_i(\phi(x) - x_{d_i}) P_{0,i}(dx) \middle| P_{0,i}(\Phi_{\nu_0, \dots, \nu_{i-2}}, \phi)^{-1} = P_{1,i} \right\} = T_i(P_{0,i}, P_{1,i}|\nu_{i-1}). \quad (2.38)$$

3 Stochastic version of Knothe-Rosenblatt type rearrangement.

Let \mathcal{A} denote the set of all \mathbf{R}^d -valued, continuous semimartingales $\{X(t)\}_{0 \leq t \leq 1}$ on a (possibly different) complete filtered probability space such that there exists a Borel measurable $\beta_X : [0, 1] \times C([0, 1]) \mapsto \mathbf{R}^d$ for which

(i) $\omega \mapsto \beta_X(t, \omega)$ is $\mathcal{B}(C([0, t]))_+$ -measurable for all $t \in [0, 1]$,

(ii) $X(t) = X(0) + \int_0^t \beta_X(s, X)ds + W_X(t)$ ($0 \leq t \leq 1$).

Here $\mathcal{B}(C([0, t]))_+ := \cap_{s>t} \mathcal{B}(C([0, s]))$, $\mathcal{B}(C([0, t]))$ and W_X denote the Borel σ -field of $C([0, t])$ and an (\mathcal{F}_t^X) -Brownian motion, respectively, and $\mathcal{F}_t^X := \sigma[X(s) : 0 \leq s \leq t]$ (see e.g. [14]). Let $d \geq 2$ and $1 \leq d_1 < d$, and let $b_1 : [0, 1] \times \mathbf{R}^{d_1} \mapsto \mathbf{R}^{d_1}$ be a Borel measurable function such that the following SDE has a weak solution for a given initial distribution:

$$dX_1(t) = b_1(t, X_1(t))dt + dW_{X_1}(t). \quad (3.1)$$

Let $L(t, x; u) : [0, 1] \times \mathbf{R}^d \times \mathbf{R}^{d-d_1} \mapsto [0, \infty)$.

A minimizer of the following can be considered as the stochastic optimal control (SOC for short) version of the Knothe Rosenblatt type rearrangement: for $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,

$$\begin{aligned} V(P_0, P_1 | b_1) &:= \inf \left\{ E \left[\int_0^1 L(t, Y(t); \beta_{Y,2}(t, Y)) dt \right] \middle| Y \in \mathcal{A}, \beta_{Y,1}(t, Y) = b_1(t, Y_1(t)), \right. \\ &\quad \left. PY(0)^{-1} = P_0, PY(1)^{-1} = P_1 \right\}, \end{aligned} \quad (3.2)$$

where we write $\beta_Y(t, Y) = (\beta_{Y,1}(t, Y), \beta_{Y,2}(t, Y)) \in \mathbf{R}^{d_1} \times \mathbf{R}^{d-d_1}$.

Example 3.1 For $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$, take T_{KR} in section 1 and, on a complete filtered probability space, consider

$$Z(t) = Z(0) + \int_0^t \frac{T_{KR}(Z(0)) - Z(s)}{1-s} ds + W_Z(t). \quad (3.3)$$

Then $Z(1) = T_{KR}(Z(0))$. In particular, $PZ(1)^{-1} = P_1$, provided $PZ(0)^{-1} = P_0$. Besides, $\beta_{Z,i}(t, Z) = \beta_{Z_i}(t, Z_i)$ for all $i = 1, \dots, d$. Suppose that $p \in [1, 2)$ and that $\int_{\mathbf{R}^d} |x|^p (P_0(dx) + P_1(dx))$ is finite. Then

$$E \left[\int_0^1 \left| \frac{T_{KR}(Z(0)) - Z(s)}{1-s} \right|^p ds \right] < \infty. \quad (3.4)$$

Indeed, $W_o(t) := Z(t) - Z(0) - (T_{KR}(Z(0)) - Z(0))t$ is a tided down brownian motion starting and arriving at 0, and

$$\frac{T_{KR}(Z(0)) - Z(s)}{1-s} = T_{KR}(Z(0)) - Z(0) - \frac{W_o(s)}{1-s}.$$

We describe our assumption in this section to show the existence of the stochastic analogue of the Knothe Rosenblatt type rearrangement.

(H.1). (i) $L \in C([0, 1] \times \mathbf{R}^d \times \mathbf{R}^{d-d_1} : [0, \infty))$, (ii) $u \mapsto L(t, x; u)$ is strictly convex.

(H.2). There exists $\gamma > 1$ such that

$$\liminf_{|u| \rightarrow \infty} \frac{\inf\{L(t, x; u) : (t, x) \in [0, 1] \times \mathbf{R}^d\}}{|u|^\gamma} > 0. \quad (3.5)$$

(H.3).

$$\Delta L(\varepsilon_1, \varepsilon_2) := \sup \frac{L(t, x; u) - L(s, y; u)}{1 + L(s, y; u)} \rightarrow 0 \quad \text{as } \varepsilon_1, \varepsilon_2 \rightarrow 0, \quad (3.6)$$

where the supremum is taken over all (t, x) and $(s, y) \in [0, 1] \times \mathbf{R}^d$ for which $|t - s| \leq \varepsilon_1$, $|x - y| < \varepsilon_2$ and over all $u \in \mathbf{R}^d$.

The following can be proved in the same way as Prop. 2.1 in [19], and the proof is omitted.

Proposition 3.1 *Suppose that (H.1)-(H.3) hold. Then for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$, $V(P_0, P_1|b_1)$ has a minimizer, provided it is finite.*

3.1 Duality Theorem

We consider the following HJB Equation:

$$\begin{aligned} \frac{\partial v(t, x)}{\partial t} + \frac{1}{2} \Delta v(t, x) + \langle \nabla_{\mathbf{x}_{d_1}} v(t, x), b_1(t, \mathbf{x}_{d_1}) \rangle \\ + H(t, x; \nabla_{\mathbf{x}_{d_1, d}} v(t, x)) = 0, \end{aligned} \quad (3.7)$$

$((t, x) \in (0, 1) \times \mathbf{R}^d)$, where

$$H(t, x; z) := \sup\{\langle u, z \rangle - L(t, x; u) | u \in \mathbf{R}^{d-d_1}\} \quad (z \in \mathbf{R}^{d-d_1}).$$

For $f \in C_b(\mathbf{R}^d)$,

$$\begin{aligned} u(t, x; f|b_1)(x) := \sup \left\{ E \left[f(Y(1)) - \int_t^1 L(s, Y(s); \beta_{Y,2}(s, Y)) ds \right] \right. \\ \left. Y(t) = x, \beta_{Y,1}(s, Y) = b_1(s, Y_1(s)), Y \in \mathcal{A} \right\}. \end{aligned} \quad (3.8)$$

(H.4). (i) $L(t, x; o)$ is bounded; (ii) $\Delta L(0, \infty)$ is finite; (iii) $b_1 \in C^{1,2}([0, 1] \times \mathbf{R}^d) \cap C_b^{0,1}([0, 1] \times \mathbf{R}^d)$, $|D_x L(t, x; u)|/(1 + L(t, x; u))$ is bounded on $[0, 1] \times \mathbf{R}^d \times \mathbf{R}^{d_1}$ and $D_u L(t, x; u)$ is bounded on $[0, 1] \times \mathbf{R}^d \times B_R$ for all $R > 0$, where $B_R := \{x \in \mathbf{R}^{d_1} | |x| \leq R\}$.

The following can be proved in the same way as Theorem 11.1 in IV.11 of [7], and the proof is omitted.

Proposition 3.2 *Suppose that (H.1)-(H.2) and (H.4,i,iii) hold. Then, for any $f \in C^5(\mathbf{R}^d) \cap C_b^3(\mathbf{R}^d)$, $u(t, x; f|b_1) \in C^{1,2}([0, 1] \times \mathbf{R}^d) \cap C_b^{0,1}([0, 1] \times \mathbf{R}^d)$ and is the unique classical solution of the HJB Equation (3.7) with $v(1, x) = f(x)$.*

It is easy to see that the following holds:

$$V(P_0, P_1|b_1) := \inf \left\{ E \left[\int_0^1 \frac{L(t, Y(t); \beta_{Y,2}(t, Y))}{1_{\{b_1(t, Y_1(t))\}}(\beta_{Y,1}(t, Y))} dt \right] \middle| Y \in \mathcal{A}, \right. \\ \left. PY(0)^{-1} = P_0, PY(1)^{-1} = P_1 \right\}, \quad (3.9)$$

which implies the duality theorem for $V(P_0, P_1|b_1)$.

Theorem 3.1 *Suppose that (H.1)-(H.4) hold. Then for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,*

$$V(P_0, P_1|b_1) = \sup \left\{ \int_{\mathbf{R}^d} v(1, y) P_1(dy) - \int_{\mathbf{R}^d} v(0, x) P_0(dx) \right\}, \quad (3.10)$$

where the supremum is taken over all classical solutions v of (3.7) with $v(1, y) \in C_b^\infty(\mathbf{R}^d)$.

(Proof). Under (H.1)-(H.3) and (H.4,i,ii), (3.9) implies that $V(P_0, \cdot|b_1)$ is convex and lower-semicontinuous, which can be proved in the same way as in [19] and is not identically equal to infinity by considering the case where $\beta_{Y,2}(s, Y) = o$ from (H.4,i). Hence, from Theorem 2.2.15 and Lemma 3.2.3 in [3],

$$V(P_0, P_1|b_1) = \sup \left\{ \int_{\mathbf{R}^d} f(y) P_1(dy) - V(P_0, \cdot|b_1)^*(f) \middle| f \in C_b(\mathbf{R}^d) \right\}, \quad (3.11)$$

where

$$V(P_0, \cdot|b_1)^*(f) := \sup \left\{ \int_{\mathbf{R}^d} f(y) P(dy) - V(P_0, P|b_1) \middle| P \in \mathcal{M}_1(\mathbf{R}^d) \right\}. \quad (3.12)$$

One can replace $C_b(\mathbf{R}^d)$ by $C_b^\infty(\mathbf{R}^d)$ in (3.11) in the same way as in the proof of Theorem 2.1 in [19]. For $f \in C_b^\infty(\mathbf{R}^d)$, from Proposition 3.2,

$$\begin{aligned}
& V(P_0, \cdot | b_1)^*(f) \\
&= \sup \left\{ E \left[f(Y(1)) - \int_0^1 \frac{L(t, Y(t); \beta_{Y,2}(t, Y))}{1_{b_1(t, Y_1(t))}(\beta_{Y,1}(t, Y))} dt \right] \middle| Y \in \mathcal{A}, P(Y(0))^{-1} = P_0 \right\} \\
&= \int_{\mathbf{R}^d} u(0, x; f | b_1) P_0(dx), \tag{3.13}
\end{aligned}$$

where the optimal control is $\beta_{Y,2}(t, Y) = \nabla_{d_1, d} u(t, Y(t); f | b_1)$. \square

As a corollary to Theorem 3.1, in the same way as [19], we easily obtain

Corollary 3.1 *Suppose that (H.1)-(H.4) hold. Then for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$ for which $V(P_0, P_1 | b_1)$ is finite, there exists a Borel measurable function $b_2^o : [0, 1] \times \mathbf{R}^d \mapsto \mathbf{R}^{d-d_1}$ such that for a minimizer $\{Y(t)\}_{0 \leq t \leq 1}$, $\beta_{Y,2}(t, Y) = b_2^o(t, Y(t))$.*

We consider the following marginal problem:

$$v(P_0, P_1 | b_1) := \inf \int_0^1 dt \int_{\mathbf{R}^d} L(t, x; B_2(t, x)) Q_t(dx), \tag{3.14}$$

where the infimum is taken over all $\{Q_t(dx)\}_{0 \leq t \leq 1} \subset \mathcal{M}_1(\mathbf{R}^d)$ for which $B_1 = b_1$, $Q_t = P_t$ ($t = 0, 1$) and

$$\frac{\partial Q_t(dx)}{\partial t} = \frac{1}{2} \Delta Q_t(dx) - \text{div}(B(t, x) Q_t(dx)),$$

in a weak sense. Here we write $B(t, x) = (B_1(t, x), B_2(t, x)) \in \mathbf{R}^{d_1} \times \mathbf{R}^{d-d_1}$.

In the same way as [17], we have

Theorem 3.2 *Suppose that (H.1)-(H.4) hold. Then for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,*

$$v(P_0, P_1 | b_1) = \sup \left\{ \int_{\mathbf{R}^d} v(1, y) P_1(dy) - \int_{\mathbf{R}^d} v(0, x) P_0(dx) \right\}, \tag{3.15}$$

where the supremum is taken over all classical solutions v of (3.7) with $v(1, y) \in C_b^\infty(\mathbf{R}^d)$. In particular, $V(P_0, P_1 | b_1) = v(P_0, P_1 | b_1) \in [0, \infty)$.

We introduce an additional assumption to formulate the duality theorem in the framework of the theory of viscosity solutions.

(H.4)'. (i) $\partial L(t, x; u)/\partial t$ and $D_x L(t, x; u)$ is bounded on $[0, 1] \times \mathbf{R}^d \times B_R$ for all $R > 0$;
(ii) $\Delta L(0, \infty)$ is finite; (iii) $b_1 \in C_b^1([0, 1] \times \mathbf{R}^d)$.

In the same way as in Lemma 4.5 in [17], one can prove

Proposition 3.3 *Suppose that (H.1)-(H.3) and (H.4)' hold. Then for any $f \in UC_b(\mathbf{R}^d)$, $u(t, x; f|b_1)$ is a bounded continuous viscosity solution of (3.7) with $v(1, x) = f(x)$ and for any $Q \in \mathcal{M}_1(\mathbf{R}^d)$ and $t \in [0, 1]$,*

$$\int_{\mathbf{R}^d} u(t, x; f|b_1) Q(dx) = \sup \left\{ E \left[f(Y(1)) - \int_t^1 L(s, Y(s); \beta_{Y,2}(s, Y)) ds \right] \right. \\ \left. PY^{-1}(t) = Q, \beta_{Y,1}(s, Y) = b_1(s, Y_1(s)), Y \in \mathcal{A} \right\}. \quad (3.16)$$

In addition, for any bounded continuous viscosity solution u of (3.7) with $u(1, x) = f(x)$, $u(t, x) \geq u(t, x; f|b_1)$, that is, $u(t, x; f|b_1)$ is minimal.

In the same way as in Theorem 3.1, from Prop. 3.3, we have

Theorem 3.3 *Suppose that (H.1)-(H.3) and (H.4)' hold. Then for any $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,*

$$V(P_0, P_1|b_1) = \sup \left\{ \int_{\mathbf{R}^d} v(1, y) P_1(dy) - \int_{\mathbf{R}^d} v(0, x) P_0(dx) \right\}, \quad (3.17)$$

where the supremum is taken over all bounded continuous viscosity solutions $v(t, x; f)$ of (3.7) with $v(1, x) \in C_b^\infty(\mathbf{R}^d)$.

Remark 3.1 *(H.3) and (i) in (H.4)' implies (i) in (H.4).*

3.2 Convergence Theorem

Let $L_1 : [0, 1] \times \mathbf{R}^{d_1} \times \mathbf{R}^{d_1} \mapsto [0, \infty)$ and $L_2 : [0, 1] \times \mathbf{R}^d \times \mathbf{R}^{d-d_1} \mapsto [0, \infty)$. For $\varepsilon > 0$, $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$,

$$V^\varepsilon(P_0, P_1) := \inf \left\{ E \left[\sum_{i=1}^2 \varepsilon^{i-1} \int_0^1 L_i(t, \mathbf{Y}_i(t); \beta_{Y,i}(t, Y)) dt \right] \right. \\ \left. PY(0)^{-1} = P_0, PY(1)^{-1} = P_1, Y \in \mathcal{A} \right\}, \quad (3.18)$$

where $\mathbf{Y}_1(t) := Y_1(t)$ and $\mathbf{Y}_2(t) := Y(t)$ for $Y(t) = (Y_1(t), Y_2(t)) \in \mathbf{R}^{d_1} \times \mathbf{R}^{d-d_1}$.

If (H.1)-(H.3) holds for $L = L_i$ for all $i = 1, 2$, then $V^\varepsilon(P_0, P_1)$ has a minimizer, provided it is finite (see Prop. 2.1 in [19]).

$$V_1(P_{0,1}, P_{1,1}) := \inf \left\{ E \left[\int_0^1 L_1(t, Y(t); \beta_Y(t, Y)) dt \right] \right. \\ \left. PY(0)^{-1} = P_{0,1}, PY(1)^{-1} = P_{1,1} \right\}, \quad (3.19)$$

where \mathcal{A}_1 denotes \mathcal{A} with $d = d_1$.

Remark 3.2 If (H.1)-(H.4) with $L = L_1$ holds and that $V_1(P_{0,1}, P_{1,1})$ is finite. Then there exists a Borel measurable function $b : [0, 1] \times \mathbf{R}^{d_1} \mapsto \mathbf{R}^{d_1}$ such that for any minimizer $\{Y(t)\}_{0 \leq t \leq 1}$ of $V_1(P_{0,1}, P_{1,1})$, $\beta_Y(t, Y) = b(t, Y(t))$ (see [19]).

Let b_1 denote the drift vector of the minimizer of $V_1(P_{0,1}, P_{1,1})$, provided it exists and let $V_2(P_0, P_1|b_1)$ denote $V(P_0, P_1|b_1)$ with $L = L_2$. Then

Theorem 3.4 Let $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$. Suppose that (H.1)-(H.3) with $L = L_i$ holds ($i = 1, 2$) and that $V_1(P_{0,1}, P_{1,1})$ and $V_2(P_0, P_1|b_1)$ is finite and have the unique minimizers $\{X_1(t)\}_{0 \leq t \leq 1}$ and $\{X(t)\}_{0 \leq t \leq 1}$, respectively. Then a minimizer $\{Y^\varepsilon(t)\}_{0 \leq t \leq 1}$ of $V^\varepsilon(P_0, P_1)$ exists and weakly converges to $\{X(t)\}_{0 \leq t \leq 1}$ as $\varepsilon \rightarrow 0$. In particular,

$$\lim_{\varepsilon \rightarrow 0} E \left[\int_0^1 L_1(t, Y_1^\varepsilon(t); \beta_{Y^\varepsilon, 1}(t, Y^\varepsilon)) dt \right] = V_1(P_{0,1}, P_{1,1}), \quad (3.20)$$

$$\lim_{\varepsilon \rightarrow 0} E \left[\int_0^1 L_2(t, Y^\varepsilon(t); \beta_{Y^\varepsilon, 2}(t, Y^\varepsilon)) dt \right] = V_2(P_0, P_1|b_1). \quad (3.21)$$

(Proof) In the same way as Prop. 2.1 in [19], one can show that there exists a minimizer $Y^\varepsilon(t)$ of $V^\varepsilon(P_0, P_1)$ since

$$V^\varepsilon(P_0, P_1) \leq V_1(P_{0,1}, P_{1,1}) + \varepsilon V_2(P_0, P_1|b_1). \quad (3.22)$$

In the same way as in Lemma 3.1 in [19], from (H.2), one can show that any sequence $\{Y^{\varepsilon_n}(\cdot)\}_{n \geq 1}$ in \mathcal{A} ($\varepsilon_n \rightarrow 0$ as $n \rightarrow \infty$) has a weakly convergent subsequence $\{Y^{\varepsilon_{n(k)}}(\cdot)\}_{k \geq 1}$. Indeed,

$$E \left[\int_0^1 L_1(t, Y_1^\varepsilon(t); \beta_{Y^\varepsilon, 1}(t, Y^\varepsilon)) dt \right] \leq V^\varepsilon(P_0, P_1), \quad (3.23)$$

$$E \left[\int_0^1 L_2(t, Y^\varepsilon(t); \beta_{Y^\varepsilon, 2}(t, Y^\varepsilon)) dt \right] \leq V_2(P_0, P_1|b_1). \quad (3.24)$$

We prove (3.24). In the same way as in Lemma 3.1 in [19], from (H.1,ii), by Jensen's inequality,

$$\begin{aligned} V_1(P_{0,1}, P_{1,1}) &\leq E \left[\int_0^1 L_1(t, Y_1^\varepsilon(t); \beta_{Y_1^\varepsilon}(t, Y_1^\varepsilon)) dt \right] \\ &\leq E \left[\int_0^1 L_1(t, Y_1^\varepsilon(t); \beta_{Y^\varepsilon, 1}(t, Y^\varepsilon)) dt \right]. \end{aligned} \quad (3.25)$$

Indeed, $Y_1^\varepsilon \in \mathcal{A}_1$ with

$$\beta_{Y_1^\varepsilon}(t, Y_1^\varepsilon) = E[\beta_{1,Y^\varepsilon}(t, Y^\varepsilon)|Y_1^\varepsilon(s), 0 \leq s \leq t]$$

(see e.g., p. 258 of [14]). (3.25) and (3.22) implies (3.24).

Let $Y^0(t)$ denote the weak limit of $\{Y^{\varepsilon_{n(k)}}(\cdot)\}_{k \geq 1}$ as $n \rightarrow \infty$. Then, again in the same way as in Lemma 3.1 in [19] and (3.25), from (H.1,ii) and (3.22)-(3.23), by Jensen's inequality,

$$\begin{aligned} V_1(P_{0,1}, P_{1,1}) &\leq E \left[\int_0^1 L_1(t, Y_1^0(t); \beta_{Y_1^0}(t, Y_1^0)) dt \right] \\ &\leq E \left[\int_0^1 L_1(t, Y_1^0(t); \beta_{Y^0,1}(t, Y^0)) dt \right] \\ &\leq \liminf_{k \rightarrow \infty} V^{\varepsilon_{n(k)}}(P_0, P_1) \leq \limsup_{k \rightarrow \infty} V^{\varepsilon_{n(k)}}(P_0, P_1) \\ &\leq V_1(P_{0,1}, P_{1,1}). \end{aligned} \tag{3.26}$$

$\beta_{Y^0,1}(t, Y^0) = \beta_{Y_1^0}(t, Y_1^0)$ from the strict convexity of L_1 in u , and Y_1^0 is equal to the minimizer X_1 of $V_1(P_{0,1}, P_{1,1})$ by the uniqueness of the minimizer of $V_1(P_{0,1}, P_{1,1})$ and we obtain (3.20). From (3.24), we also have

$$\begin{aligned} V_2(P_0, P_1|b_1) &\leq E \left[\int_0^1 L_2(t, Y^0(t); \beta_{Y^0,2}(t, Y^0)) dt \right] \\ &\leq \liminf_{k \rightarrow \infty} E \left[\int_0^1 L_2(t, Y^{\varepsilon_{n(k)}}(t); \beta_{Y^{\varepsilon_{n(k)},2}}(t, Y^{\varepsilon_{n(k)}})) dt \right] \\ &\leq \limsup_{k \rightarrow \infty} E \left[\int_0^1 L_2(t, Y^{\varepsilon_{n(k)}}(t); \beta_{Y^{\varepsilon_{n(k)},2}}(t, Y^{\varepsilon_{n(k)}})) dt \right] \\ &\leq V_2(P_0, P_1|b_1). \end{aligned} \tag{3.27}$$

The uniqueness of the minimizer of $V_2(P_0, P_1|b_1)$ completes the proof. \square

One can easily prove

Corollary 3.2 *Let $P_0, P_1 \in \mathcal{M}_1(\mathbf{R}^d)$. Suppose that (H.1)-(H.3) with $L = L_i$ holds ($i = 1, 2$), that $\gamma = 2$ in (H.2), and that $V_1(P_{0,1}, P_{1,1})$ and $V_2(P_0, P_1|b_1)$ is finite. Then the minimizers $\{X_1(t)\}_{0 \leq t \leq 1}$, $\{X(t)\}_{0 \leq t \leq 1}$ and $\{Y^\varepsilon(t)\}_{0 \leq t \leq 1}$ of $V_1(P_{0,1}, P_{1,1})$, $V_2(P_0, P_1|b_1)$ and $V^\varepsilon(P_0, P_1)$ exist uniquely, respectively. In addition, $\{Y^\varepsilon(t)\}_{0 \leq t \leq 1}$ weakly converges to $\{X(t)\}_{0 \leq t \leq 1}$ as $\varepsilon \rightarrow 0$ and (3.20)-(3.21) holds.*

From (3.21)-(3.22) and (3.25), we easily have

Proposition 3.4 *Suppose that the assumption in Theorem 3.3 holds. Then for any minimizer $\{Y^\varepsilon\}_{0 \leq t \leq 1}$ of $V^\varepsilon(P_0, P_1)$,*

$$0 \leq \frac{E \left[\int_0^1 L_1(t, Y_1^\varepsilon(t); \beta_{Y^\varepsilon, 1}(t, Y^\varepsilon)) dt \right] - V_1(P_{0,1}, P_{1,1})}{\varepsilon} \rightarrow 0 \quad (\varepsilon \rightarrow 0). \quad (3.28)$$

We don't know the real convergence rate!

4 Discussion

In section 2, Theorem 2.3, we assumed that $\nu \mapsto T_i(P_{0,i}, P_{1,i}|\nu)$ is continuous ($i = 3, \dots, k$). This continuity is known only in the case of the Knothe-Rosenblatt rearrangement where the representation of the minimizer is known. It is difficult to prove that $\nu \mapsto T(P_0, P_1|\nu)$ is continuous, which is our future problem.

In section 3.2, we only considered the case where $k = 2$ because of the similar reason to above. The point is that we do not even know any example such as the Knothe-Rosenblatt rearrangement. This is also our future problem.

The Knothe-Rosenblatt rearrangement implies the Brunn-Minkowskii inequality. We would like to find, in future, the inequality which can be obtained by the result in section 3.

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